# IM-UFRJ Probability Webinar Scaling Limit for Heavy-tailed Ballistic Deposition with p-sticking



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#### Ballistic Deposition (with p-sticking)

Imagine we have a cluster of blocks growing above the real line which evolves as follows:



• Whenever a block falls down on x, it has 2 possibilities: () with prob. p [0,1], it sticks to the first point of contact.

② with prob. 1-p, it falls straight down and sticks to the last block above x.





p=1 (standard Ballistic Deposition)

p=0 (Random Deposition)



Given x E Z and T>O, let us define h(x,T) := height at time T of the column of blocks growing above x.

Then:

- · For p=0 (RANDOM DEPOSITION):
  - · h(x,T) grows linearly in T
  - h(n, T) has Gaussian fluctuations
- · For p=1 (BALLISTIC DEPOSITION):
  - h(2,T) grows linearly in T
  - h(n, T) is conjectured to have KPZ fluctuations.

<u>Question</u> How does the system transition from one behavior to the other as p goes from 1 to 0?

- Answer: Too hard for us at the moment.
- lolea: ask the same question in a different model.

#### Heavy-tailed Ballistic Deposition (with p-sticking)

The model is the same as BD, with the only difference that the blocks have now <u>random heights</u>, which are i.i.d. with a common CDF F, satisfying:

- F is continuous
- F has a regularly varying (right) tail with index of (0,2), i.e.

I - F(t) = t L(t) for some slowly varying function L.

**Problem:** Understand the rate of growth of the cluster of blocks as  $T \rightarrow \infty$ .



[BHKSV9]] = Buldyrev, Harlin, Kertész, Vicsek (Ballistic Deposition with power-law noise)



#### Heavy-Tailed Ballictic Deposition



#### Standard Ballistic Deposition



Heavy-Tailed Random Deposition

## Main results

Let h(x,T;p) denote the height above  $x \in \mathbb{Z}$  of the cluster at time T for the model with p-sticking. Assume that  $h(\cdot,0;p) \equiv 0$ .

Theorem A There exists a slowly varying function  $L_{F}$  such that for any pro and  $x \in \mathbb{Z}$ ,  $\frac{h(x,T;P)}{(pT^{2})^{\frac{1}{4}}L_{F}(pT^{2})} \xrightarrow{d} \mathcal{H}(x)$ 

For p=o, the behavior is different : the height process (h(n,T;o))\_T>o is a compound Poisson process with jump CDF F. Thus, by the generalized CLT,

$$\frac{h(x,T;o) - c_T(\alpha)}{T \bigcup_F (T)} \xrightarrow{d} \text{stable law of index al}}$$

In particular, h(x,T;o) << T<sup>2/d</sup> L<sub>F</sub>(T<sup>2</sup>).

Question: What can we say about the phase transition as p > 0 ?

**Theorem B** Fix  $\delta \in (0,1)$  and, for each T>0, set  $P_T := T^{-\delta}$ . Then the system  $(h(\alpha,T;P_T))_{T>0}$  exhibits a phase transition at  $\mathcal{V}_{c} = (2-d)A1$ : (i)  $\delta < \mathcal{V}_{c} \implies \frac{h(\alpha,T;P_T)}{(P_T^{T^2})^{\frac{1}{\alpha}}L_{\mu}(P_T^{T})} \xrightarrow{d} \mathcal{I}_{c}$ (i)  $\delta > \mathcal{V}_{c} \implies \frac{h(\alpha,T;P_T)}{(P_T^{T^2})^{\frac{1}{\alpha}}L_{\mu}(P_T^{T})} \xrightarrow{IP} \infty$  if  $\alpha \in (1,2)$  (and is not tight)

$$\mathbf{x}_{c} \implies \underbrace{\mathsf{n}(\mathbf{x}_{1},1;\mathbf{p}_{T})}_{(\mathbf{p}_{T}^{-2})^{\frac{1}{n}} \mathsf{L}_{\mu}(\mathbf{p}_{T}^{-1})} \xrightarrow{\mathsf{IP}} \infty \quad \text{if } a \in (1,2) \quad (and is not +i)$$

# The limit H

Let  $\mathcal{L} := \int S : [0,1] \longrightarrow |R : S(1) = 0$ ,  $|S(t_1) - S(t_2)| \leq |t_1 - t_2|, t_1, t_2 \in [0,1]$ = f 1 - Lipschitz paths ending at 0 3

and  $\Delta := \{(t,x) \in \mathbb{R}^3 : |x| \le 1 - t\} = triangle of vertices (0,1), (0,-1), (1,0).$ (0,1) Note that graph (s)  $\subseteq \Delta$  for any s  $\in \mathcal{L}$ .



turthermore, let:

- $(U_n)_{n \in \mathbb{N}}$  be i.i.d. random variables uniformly distributed on  $\Delta$
- · Let O< X1 < X2 < ... be the sequence of points in a Poisson process with rate  $\pm$  independent of  $(U_n)_{n \in \mathbb{N}}$  and define  $M_n := \frac{\pm}{(X_n)_{X_n}}$ . Observe that (Mn)n GIN is decreasing , i.e. M1 > M2 > ... > 0.
- · We think of  $M_n$  as a (positive) weight positioned at  $U_n \in \Delta$  .
- Then, we define  $\mathcal{H} := \max \left\{ \sum_{n \in \mathbb{N}} M_n \perp \left\{ U_n \in \operatorname{graph}(s) \right\} : s \in \mathcal{L} \right\}$ = maximum sum of weights Mn that can be collected by any 1-Lipschitz path ending at 0.

It can be seen that the max is always achieved by an a.s- unique such path.

### Relationship with Last Passage Percolation

For simplicity, let us take p=1.

Consider the following "graphical construction" of the process:



- To each z ∈ Z, assign an independent (marked) Poisson process with rate 1.
- · Each x in the x-th column indicates an occurrence of the corresponding process
- X=(t,x) means that a block fell down above site x and stuck to the cluster precisely at time t
- Each x= (t,x) comes with an independent "weight" η(t,x) which represents the height of the block.
- A (discrete) path S: [0,7] → Z is called admissible if:
  - i) it is cadlag (⇒ piecewise constant)
  - ii) all jumps of s are nearest-neighbour (i.e.  $\Delta S(t) := S(t) S(t^{-}) \in \{-1, 0, 1\} \ \forall t)$
- iii) 5 can only jump if itlands on a 🗙 (i.e. ∆s(t)≠0 =D (t,s(t))=X)
- Define (A<sub>T</sub>(x):= f S: [o,T] → Z: S admissible and S(T) = x }.

Proposition For any T>O and x & Z, almost surely we have

$$h(x,T; 1) = \max \left\{ h(s(0), 0) + \sum_{\Delta S(t) \neq 0} \eta(t, s(t)) : s \in \mathcal{A}_{T}(x) \right\}$$

Thus, to prove our results, we need to show that this "discrete" optimization problem above is asymptotically equivalent to the "continuous" one defining  $\mathcal{H}$  as  $T \longrightarrow \infty$ .

A few words about the proof
The limit *M* is the "same" one obtained by Hambly & Martin (2007)
for Heavy-tailed Last Passage Percolation on IN × IN. More precisely, if:

(*n*<sub>i,j</sub>)<sub>i,j ∈ IN</sub> are *i*·*i*.d. with common CDF F as before,
H<sub>n</sub> := maximum sum of weights *n*<sub>i,j</sub> that can be collected by
any increasing path (0,0) → (n,n),

then

$$\frac{H_n}{n^{2/4} L_{F}(n)} \xrightarrow{J} \widetilde{\mathcal{H}}$$

· To prove our results, we will follow the approach by Hambly & Martin (2007).

#### · Difficulties

- We don't have a regular lattice like IN×IN, but rather a random "Roissonian lattice". In particular, admissible paths are harder to control than increasing paths.
   (discrete) admissible paths ≈ 1-Lipschitz paths <u>only</u> as T→∞.
   Hambly & Martin only considered the "p=1" case.
- <u>MAIN IDEAS</u>: How to show that  $\frac{h(0,T;p)}{(pT^2)^4 L_p(pT^2)} \approx \mathcal{H}$ . •  $\mathcal{C}_T = \int X = (4,\infty)$ : X can be "collected" by some  $s \in \mathcal{H}_T(0)$
- Order the weights  $\{\eta(t,x) : (t,x) \in \mathcal{E}_T\}$  by size, obtaining a decreasing sequence  $M_1^{(T)} > M_2^{(T)} > \dots > M_k^{(T)}$ .
- Let  $U_n^{(T)}$ ,  $n=1,\ldots,k_T$ , denote the position of the weight  $M_n^{(T)}$ .
- Call  $(U_n^{(\tau)}, M_n^{(\tau)}: n=1, \dots, k_T)$  the discrete model.
- · Call (Un, Mn) the continuous model.

Then, the main steps of the proofs are the following: 1) Show that only the largest weights matter, i.e. for K large,

$$h(0,T;p) \approx \max \left\{ \sum_{n \leq K} M_n^{(T)} \coprod \{ U_n^{(T)} \in \operatorname{graph}(s) \} : s \in \mathcal{A}_T^{(0)} \right\} =: S_K^{(T)}$$

tor this, we need good bounds on:  
i) 
$$|E(M_n^{(T)})$$
 for all n,  
ii)  $|E(\max_{s \in A_T^{(0)}} | f i \leq n : \widetilde{M}_i \text{ is collected by } s \mathcal{J} |$ 

2) Show that the same holds for the continuous model, i.e.  
for all K large enough,  
$$\mathcal{H} \approx \max \left\{ \sum_{n \leq K} M_n \operatorname{IL}_{\{U_n \in \operatorname{graph}(s)\}} : s \in \mathcal{L} \right\} =: \mathcal{S}_K$$

3) Construct a coupling between  $(U_n^{(T)}, M_n^{(T)})_{n \in K_T}$  and  $(U_n, M_n)_{n \in W}$  such that: i) for each  $K \in W$ ,  $\frac{1}{(pT^2)^{\frac{1}{4}}L_r(T)} (M_1^{(T)}, M_K^{(T)}) \xrightarrow{c.s.}{T \to \infty} (M_1, \dots, M_K)$ 

(i) for each i, with high probability as  $T \rightarrow \infty$ ,

The optimal path in the discrete model (T) The optimal path in the continuous model collects the weight  $M_i^{(T)}$  (T) collects the weight  $M_i$ 

(i) + (ii) imply that , for each K, if T is large enough ,

$$\frac{1}{(p\tau^2)^{l_{\alpha}}L_{p}(\tau)} S_{\kappa}^{(\tau)} \approx S_{\kappa}$$

which, combined with (1) and (2), gives the result.